```
import numpy as np
import tensorflow as tf
import keras
import struct
from array import array
from keras._tf_keras.keras import datasets, layers, models
from os.path import join
import matplotlib.pyplot as plt
# Define file paths for MNIST data files
training_images_filepath = '/content/train-images.idx3-ubyte
training_labels_filepath = '/content/train-labels.idx1-ubyte
test_images_filepath = '/content/t10k-images.idx3-ubyte
test_labels_filepath = '/content/t10k-labels.idx1-ubyte'
# Define the MnistDataloader class (as provided)
class MnistDataloader(object):
    def __init__(self, training_images_filepath, training_labels_filepath,
                  test_images_filepath, test_labels_filepath):
        self.training_images_filepath = training_images_filepath
        self.training_labels_filepath = training_labels_filepath
        self.test_images_filepath = test_images_filepath
        self.test_labels_filepath = test_labels_filepath
    def read_images_labels(self, images_filepath, labels_filepath):
        labels = []
        with open(labels_filepath, 'rb') as file:
            magic, size = struct.unpack(">II", file.read(8))
            if magic != 2049:
                 raise ValueError('Magic number mismatch, expected 2049, got {}'.format(magic))
            labels = array("B", file.read())
        with open(images_filepath, 'rb') as file:
            magic, size, rows, cols = struct.unpack(">IIII", file.read(16))
            if magic != 2051:
                raise ValueError('Magic number mismatch, expected 2051, got {}'.format(magic))
            image_data = array("B", file.read())
        images = []
        for i in range(size):
            images.append([0] * rows * cols)
        for i in range(size):
            img = np.array(image_data[i * rows * cols:(i + 1) * rows * cols])
            img = img.reshape(28, 28)
            images[i][:] = img
        return images, labels
    def load_data(self):
        x\_train, \ y\_train = self.read\_images\_labels(self.training\_images\_filepath), \ self.training\_labels\_filepath)
        x_test, y_test = self.read_images_labels(self.test_images_filepath, self.test_labels_filepath)
        return (x_train, y_train), (x_test, y_test)
# Instantiate the dataloader and load the data
mnist_dataloader = MnistDataloader(training_images_filepath, training_labels_filepath, test_images_filepath, test_labels_filepath)
(x_{train}, y_{train}), (x_{test}, y_{test}) = mnist_dataloader.load_data()
# Convert to numpy arrays
x_train = np.array(x_train)
x_test = np.array(x_test)
y_train = np.array(y_train)
y_{test} = np.array(y_{test})
# Reshape data to add a single channel dimension (grayscale)
x_train = x_train.reshape((x_train.shape[0], 28, 28, 1))
x_{\text{test}} = x_{\text{test.reshape}}((x_{\text{test.shape}}[0], 28, 28, 1))
\# Normalize pixel values between 0 and 1
x_{train} = x_{train.astype('float32')} / 255.0
x_{test} = x_{test.astype('float32')} / 255.0
model = models.Sequential([
    layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)),
    layers.MaxPooling2D((2, 2)),
    layers.Conv2D(64, (3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),
    layers.Flatten(),
    layers.Dense(64, activation='relu'),
    layers.Dense(10, activation='softmax')
1)
```

/usr/local/lib/python3.10/dist-packages/keras/src/layers/convolutional/base\_conv.py:107: UserWarning: Do not pass an `input\_shape`/ super().\_\_init\_\_(activity\_regularizer=activity\_regularizer, \*\*kwargs)

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	0
conv2d_1 (Conv2D)	(None, 11, 11, 64)	18,496
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	0
flatten (Flatten)	(None, 1600)	0
dense (Dense)	(None, 64)	102,464
dense_1 (Dense)	(None, 10)	650

Total params: 121,930 (476.29 KB) Trainable params: 121,930 (476.29 KB)

```
model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
history = model.fit(x_train, y_train, epochs=50,validation_split=0.1)
→ Epoch 1/50
     1688/1688
                                  — 48s 27ms/step - accuracy: 0.9021 - loss: 0.3254 - val_accuracy: 0.9812 - val_loss: 0.0590
     Epoch 2/50
test_loss, test_acc = model.evaluate(x_test, y_test, verbose=2)
print(f'\nTest accuracy: {test_acc}')
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Model Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend(loc='upper left')
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Train Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Model Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend(loc='upper left')
plt.show()
predictions = model.predict(x_test)
# Example: Print the predicted class for the first test image
print("Predicted \ class \ for \ the \ first \ test \ image:", \ predictions[0].argmax())
print("Actual class for the first test image:", y_test[0])
```